

## Optimal gating angles for a three-phase 60 Hz voltage source multi-level inverter based on intelligent algorithms

Taha Ahmad Hussein<sup>1</sup>, Dahaman Ishak<sup>2</sup>, Jawad Hasan Alkhateeb<sup>3</sup>, Mohamed Salem<sup>4</sup>

<sup>1</sup>Department of Electrical Engineering Techniques, Northern Technical University, USM Offshore Center, İstanbul, Türkiye

<sup>2</sup>Department of Electrical Engineering, Prince Mohammad Bin Fahd University, Al Khobar, Saudi Arabia

<sup>3</sup>Department of Computer Engineering, Prince Mohammad Bin Fahd University, Al Khobar, Saudi Arabia

<sup>4</sup>School of Electrical and Electronic Engineering, Universiti Sains Malaysia, Penang, Malaysia

### Article Info

#### Article history:

Received Nov 16, 2024

Revised Jan 28, 2025

Accepted Mar 1, 2025

#### Keywords:

Genetic algorithm

Gray wolf optimization

Multi-level inverter

Particle swarm optimization

Slime mould algorithm

Total harmonic distortion

Whale optimization algorithm

### ABSTRACT

The three-phase multi-level inverter is considered one of the main power sources in industrial applications as well as in renewable energy applications. Therefore, researchers are interested in improving the efficiency of the inverter by reducing the total harmonic distortion (THD) value to its lowest limits. Also, one of the factors for improving the efficiency of the inverter is reducing the number of switches used, as it contributes to reducing the resulting losses. This research resorts to using many optimal algorithms to find the optimal values for the inverter gating angles that ensure reducing the THD value, as well as using a suitable topology with a least number of switches. The research used five algorithms known for their accuracy and efficiency, genetic algorithm (GA), gray wolf optimization (GWO), particle swarm optimization (PSO), slime mould algorithm (SMA), and whale optimization algorithm (WOA) separately. Then, extracting the distinctive characteristics of these algorithms in a hybrid curve and using it in driving the three-phase multi-level inverter (MLI) with 31 levels. The research displays the voltages and currents of the inverter as well as the frequency analysis for three-phase inductive load resulting from simulating the inverter using MATLAB software.

*This is an open access article under the [CC BY-SA](#) license.*



### Corresponding Author:

Taha Ahmad Hussein

Department of Electrical Engineering Techniques, Northern Technical University, USM Offshore Center  
İstanbul, Türkiye

Email: taha\_hussian@yahoo.com

## 1. INTRODUCTION

Multi-level inverter (MLI) is widely used within high power and medium voltages, and this requires a combination of several factors to obtain high efficiency for the system [1]. First, numerous DC sources are necessary to generate the required multiple levels in the output waveform. Therefore, the challenge is to use the least number of these DC sources and simultaneously securing the required voltage levels [2]. The output waveform's multiple levels are accomplished by adding or subtracting these DC sources from each other at each step. The second factor relates to the number of power switches used. These switches provide paths for the electrical currents through continuous opening and closing. The fewer the switches, the less electrical power is wasted, which reduces the total loss of the inverter and contributes effectively to higher efficiency. The most critical third factor is selecting the optimal switching times for the operation of the switches [3]. Hence, this requires complex algorithms to find the optimal switching times to operate these power switches. Optimal timing effectively contributes to diminishing the total harmonic distortion (THD) in the inverter output waveform. Lower THD is important to comply with the grid standards. The race is now on how to find the best

multi-level inverter circuit topology that has the fewest DC sources and the fewest number of switches. After that, the optimal switching time to operate the power switches can be obtained by using intelligent algorithms [4]. The algorithms can contribute to reducing the total harmonic distortion in the output waveform MLI. The output waveform consists of staircase-like levels, and when analyzed further with the Fourier series, it results in a set of trigonometric harmonic equations accompanying the fundamental harmonic. The inverter switching times depend on the modulation index. To find the optimal switching times, intelligent algorithms can be used. These algorithms result in different THD values associated with a particular value of (m). This paper aims to address the above-mentioned research problems currently found in three-phase MLIs i.e., to deploy the minimal number of DC sources, to use the minimum number of power switches, and to apply intelligent algorithms to get minimum THDs for a broad spectrum of modulation indices. This research concentrates on designing a 3-phase MLI composed of thirty-one levels and optimal gating angles derived from genetic algorithms (GA), gray wolf optimization (GWO), particle swarm optimization (PSO), slime mould algorithm (SMA), and whale optimization algorithm (WOA). Ultimately, the research studies indicate that the stepped wave MLI is the best way to obtain reduced THD values, and the success lies in using the best methods to control the operation of the inverters. The stages of operation for different control methods are presented and the solution of the trigonometric equations accompanied with stepped wave output of MLI are explained.

## 2. METHOD

This section describes the design and implementation of a 3-level three-phase MLI used as a three-phase voltage source with a minimum number of DC sources and switches, reducing the power losses and enhancing efficiency. The control method relies on selective harmonic elimination based on intelligent algorithms to achieve minimum percentage values of THD. The implementation of 31-level three-phase inverter for MATLAB simulation starts with writing the trigonometric equations for selective (targeted) harmonic elimination (SHE) to get the optimal solution of the optimal angles using the five selected algorithms, i.e. GA, GWO, PSO, SMA, and WOA. The next step is to select the optimal angles from the appropriate algorithm based on the hybrid optimization curve. These optimal angles are used as gate signals to power switches to simulate the 31-level three-phase inverter.

### 2.1. Three-phase 31-level MLI topology

The first stage includes designing a basic unit for single-phase MLI as shown in Figure 1. In general, the basic unit can provide  $2^n - 1$  levels, where n is the number of DC sources. For instance, if  $n=2$ , then the basic unit provides  $2^2-1 = 3$  levels for the positive half cycle of the inverter output voltage, and similarly another 3 levels for the negative half cycle, hence a total of 7 levels for one electrical cycle (the zero level is repeated) as shown in Figure 2.

The basic unit can be extended to provide any required level. For example, to get 15 levels, two basic units are put together with 4 DC sources, i.e.  $2^4 - 1=15$  levels per half cycle are generated as depicted in Figure 3, making 31 levels in one complete cycle. The output voltage waveform has a stepped waveform, and each step represents the value of  $V_{DC}$ . To obtain a 31-level voltage waveform from a single-phase MLI, the magnitude of the DC sources is set as  $V_{DC}$ ,  $2 V_{DC}$ ,  $5 V_{DC}$ , and  $10 V_{DC}$ , respectively, as shown in Figure 4. The process of addition and subtraction of these DC sources will result in 31 levels, including the zero level of the voltage output from the inverter [5].

The theory of Fourier series for a stepped periodic function can be presented by (1) [6].

$$v_o(t) = \frac{a_o}{2} + \sum_{k=0}^n a_n \cos(nwt) + b_n \sin(nwt) \quad (1)$$

Where  $a_o$ , represents the average value of the output voltage,  $a_n$  and  $b_n$ , are even and odd parameters of the staircase periodic signal, respectively. The stepped waveform exhibits a quarter wave symmetry which establishes  $a_o$ ,  $a_n$  and the even  $b_n$  values to zero and modifies in (1) to (2).

$$v_o(t) = \sum_{i=1,3,5,7,9,11,.....}^n b_n \sin(nwt) \quad (2)$$

Where  $b_n$  is found to be (3).

$$b_n = \frac{4V_{dc}}{n\pi} \sum_{k=1,3,5,7,9,11,.....}^m \cos(n\Phi_k) \quad (3)$$

In three-phase systems, the triplen harmonics ( $3^{rd}$ ,  $9^{th}$ ,  $15^{th}$ ,  $21^{st}$ , and  $27^{th}$ ) are eliminated implicitly. The (3) is to be solved for the optimized 15 angles of 31-level MLI to eliminate the explicit 14 harmonics ( $5^{th}$ ,  $7^{th}$ ,  $11^{th}$ ,  $17^{th}$ ,  $19^{th}$ ,  $23^{rd}$ ,  $25^{th}$ ,  $29^{th}$ ,  $31^{st}$ ,  $35^{th}$ ,  $37^{th}$ ,  $41^{st}$ , and  $43^{rd}$ ).

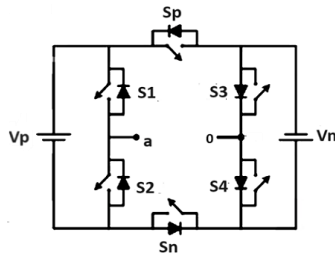


Figure 1. Basic unit of the inverter

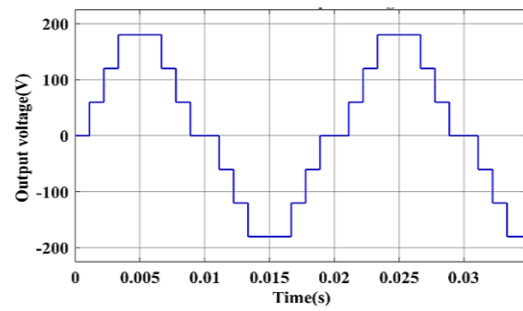


Figure 2. 7-level voltage waveform

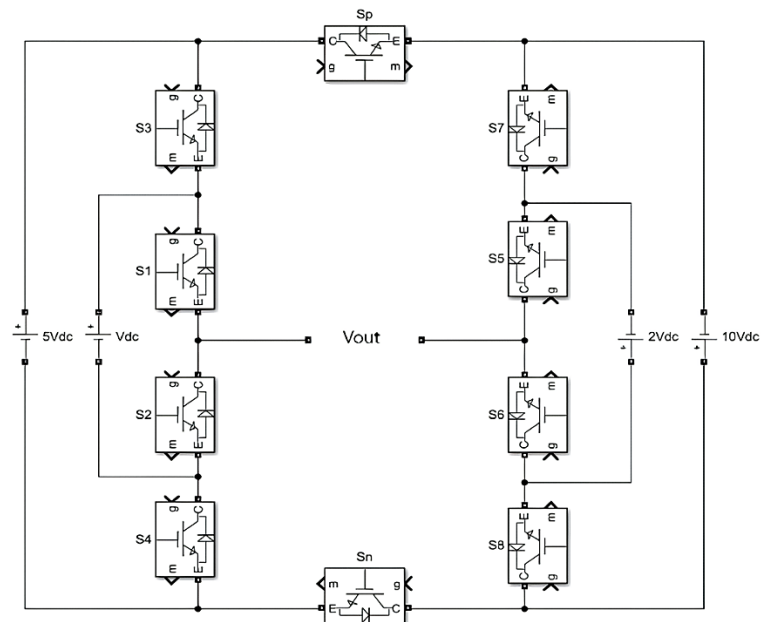


Figure 3. 31-level single-phase inverter

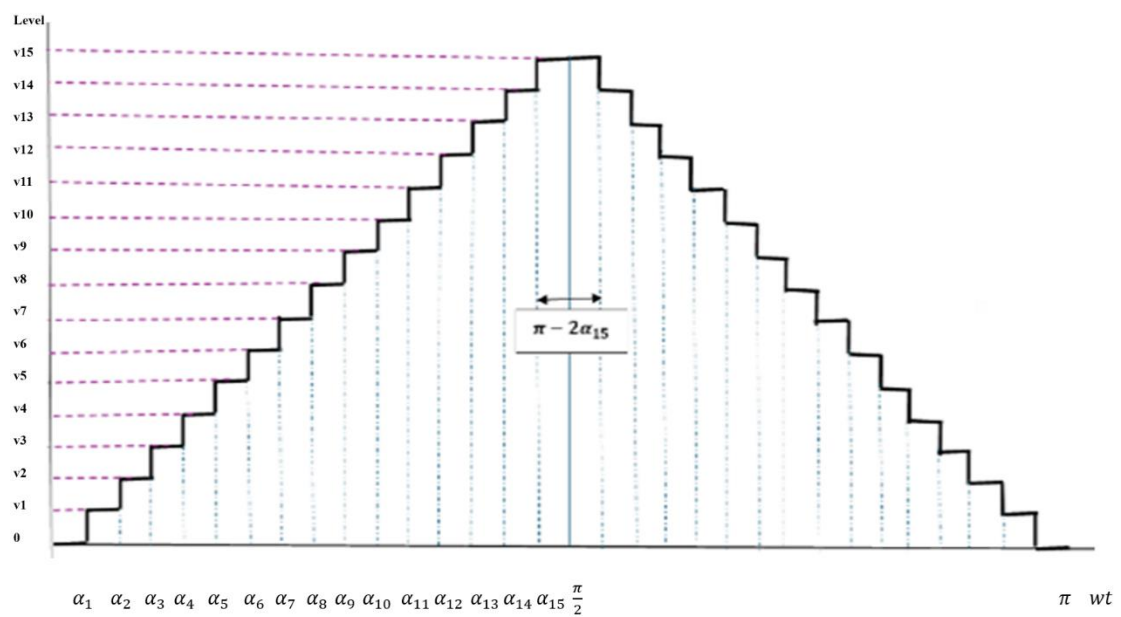


Figure 4. 31-level stepped waveform

## 2.2. Optimization techniques

Five algorithms known for their efficiency and widely used in practical applications in the multi-level inverter field, namely genetic algorithm (GA) [7]-[11], gray wolf optimization (GWO) [12]-[16], particle swarm optimization (PSO) [17]-[22], slime mould algorithm (SMA) [23]-[28], and whale optimization algorithm (WOA) [29]-[33] are calculated for finding the optimal gating angles of the power switches for whole range of modulation index ( $m$ ),  $0 < m < 1$ . Figure 5 shows the MATLAB calculated optimized angles for GA, while THD% vs modulation index for GA is shown in Figure 6. THD% is between 4% to 13%, approximately, depending on the value of the modulation index. Figure 7 shows the MATLAB calculated optimized angles for GWO. THD% vs modulation index for GWO is shown in Figure 8. THD% is between 2% to 24%, approximately, depending on the value of the modulation index. MATLAB calculated optimized angles for PSO are shown in Figure 9, while Figure 10 shows THD% vs modulation index for PSO. THD% is between 3% to 22%, approximately, depending on the value of the modulation index.

The optimized angles calculated by MATLAB for SMA are shown in Figure 11. THD% vs modulation index for SMA is shown in Figure 12. THD% is between 2% to 28%, approximately, depending on the value of the modulation index. Figure 13 shows the MATLAB calculated optimized angles for WOA and Figure 14 shows THD% vs modulation index for WOA. THD% is between 2.5% to 14%, approximately, depending on the value of the modulation index. An optimal hybrid curve (OHC) is derived from the five algorithms, i.e. GA, GWO, PSO, SMA, and WOA by taking the lowest value of the THD% for a specific value of modulation index as shown in Figure 15. This optimal relationship is employed to calculate the fifteen angles corresponding to every value of the modulation index.

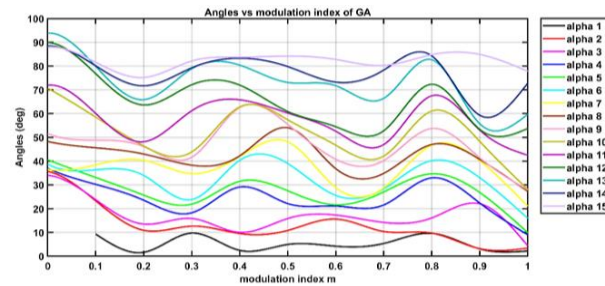


Figure 5. Angles vs  $m$  for GA

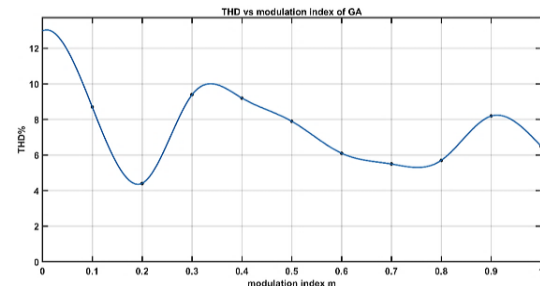


Figure 6. THD% vs  $m$  for GA

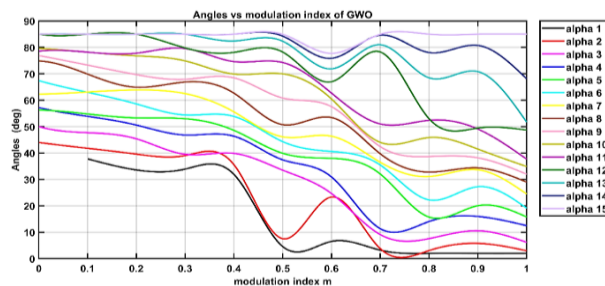


Figure 7. Angles vs  $m$  for GWO

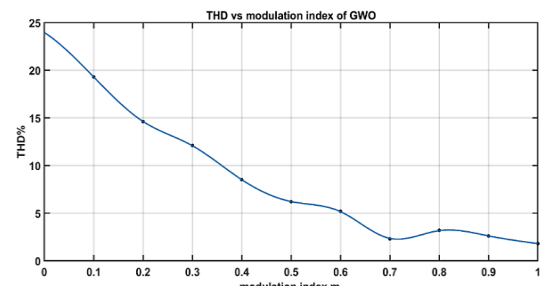


Figure 8. THD% vs  $m$  for GWO

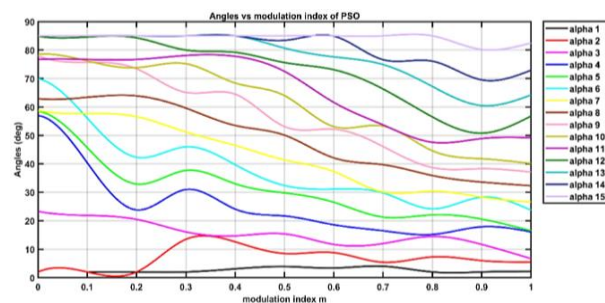


Figure 9. Angles vs  $m$  for PSO

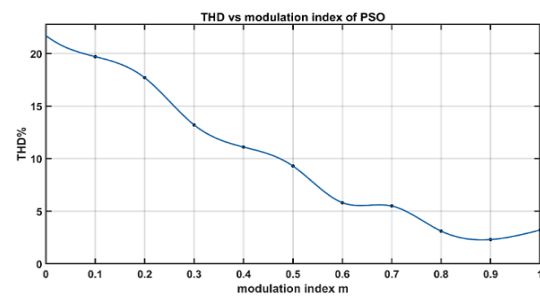


Figure 10. THD% vs  $m$  for PSO

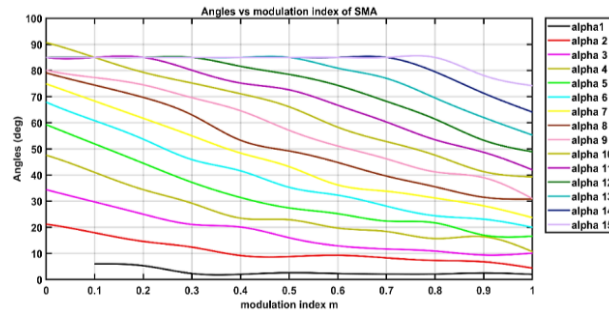


Figure 11. Angles vs m for SMA

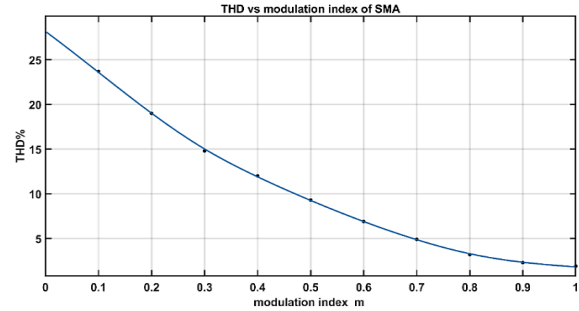


Figure 12. THD% vs m for SMA

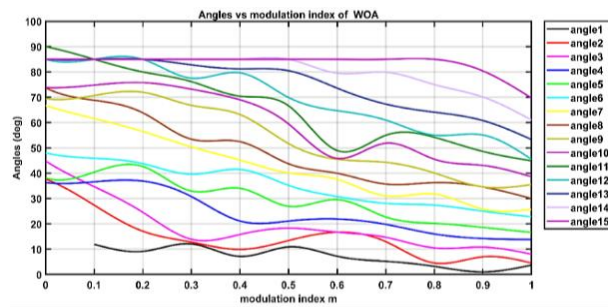


Figure 13. Angles vs m for WOA

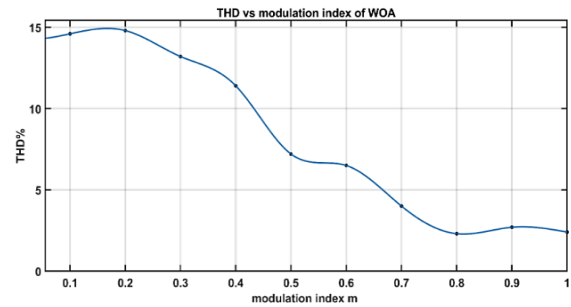


Figure 14. THD% vs m for WOA

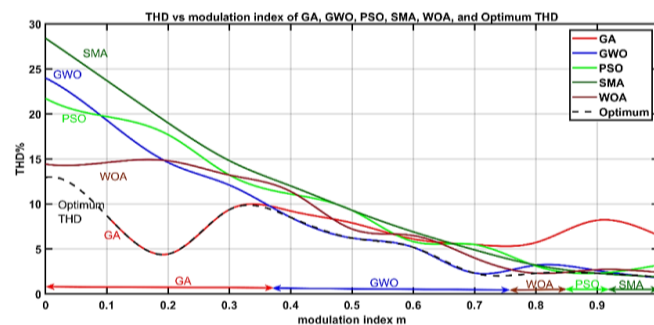


Figure 15. Optimal hybrid curve derived from GA, GWO, PSO, SMA, and WOA vs THD%

According to the chosen modulation index the user can index the appropriate algorithm from the OHC which assures the lowest THD% for the given modulation index as tabulated in Table 1. One of the five algorithms can be adopted according to the range of  $m$  to be used as depicted in Table 2. This is to ensure that minimum THD% is always achievable at the MLI output.

Table 1. THD% values achieved using OHC versus individual algorithms

Algorithm	Modulation index									
	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1
GA	9	4.2	9.8	9.7	8	6	5.8	5.9	8.1	6.8
GWO	19	14.9	12.5	9	6	5	2.4	3.8	2.8	2.3
PSO	19.8	17.2	13.8	10.9	9.8	5.9	5.3	3.9	2.5	3.8
SMA	24	19	15	12.5	9.8	7.5	5	3.8	1.2	0.9
WOA	14.8	15	13.8	11.9	7.2	6.1	4.8	2.5	3	2.6
OHC	9	4.2	9.8	9	6	5	2.4	2.5	1.2	0.9

Table 2. Modulation index range vs algorithm used

Modulation index range	Algorithm used	Modulation index range	Algorithm used
0-0.38	GA	0.86-0.92	PSO
0.39-0.75	GWO	0.93 -1	SMA
0.76-0.85	WOA		

### 3. RESULTS AND DISCUSSION

#### 3.1. MATLAB Simulink simulation

To obtain the output voltage of the three-phase inverter with 31 levels, Table 3 is prepared, which represents the state of the switches for the positive half-cycle. For the negative half-cycle, a phase difference of  $180^\circ$  is added. The required tables for the three phases are prepared, considering the  $120^\circ$  difference between the three phases.

A proposed three-phase 31 multi-level inverter (MLI) is star connected as shown in Figure 16 with 12 DC sources and 30 IGBT's ensuring the least possible loss of power resulting from the on-off operation of these switches. The three-phase inverter is simulated in the MATLAB program and the Simulink model for the switches gating of the positive half cycle of phase A is shown in Figure 17. Figure 18 represents the  $120^\circ$  phase difference between the three phases, while Figure 19 shows the gating angles for the S1, S3, S5, S7, and S9 switches for phase A, which represents the complements of S2, S4, S6, S8, and S10, respectively. These switches operate with different frequencies according to Table 3. To simulate a 60 Hz power source with phase voltage  $V_{rms} = 120$  V, these DC voltage sources are used:  $V_{DC} = 12$  V,  $2 V_{DC} = 24$  V,  $5 V_{DC} = 60$  V, and  $10 V_{DC} = 120$  V.

Table 3. Phase A switches states for the positive cycle

S1 S2 *	S3 S4 *	S5 S6 *	S7 S8 *	S9 S10 *	Level	Duration	DC sources
1	1	1	1	1	0	$\Phi_1$	No DC source
0	1	1	1	1	1	$\Phi_2 - \Phi_1$	$V_1$
1	1	0	1	1	2	$\Phi_3 - \Phi_2$	$V_2$
0	1	0	1	1	3	$\Phi_4 - \Phi_3$	$V_1 + V_2$
1	0	1	1	1	4	$\Phi_5 - \Phi_4$	$V_5 - V_1$
0	0	1	1	1	5	$\Phi_6 - \Phi_5$	$V_5$
1	0	0	1	1	6	$\Phi_7 - \Phi_6$	$V_5 + V_2 - V_1$
0	0	0	1	1	7	$\Phi_8 - \Phi_7$	$V_5 + V_2$
1	1	1	0	1	8	$\Phi_9 - \Phi_8$	$V_{10} - V_2$
0	1	1	0	1	9	$\Phi_{10} - \Phi_9$	$V_{10} - V_2 + V_1$
1	1	0	0	1	10	$\Phi_{11} - \Phi_{10}$	$V_{10}$
0	1	0	0	1	11	$\Phi_{12} - \Phi_{11}$	$V_{10} + V_1$
1	0	1	0	1	12	$\Phi_{13} - \Phi_{12}$	$V_{10} - V_2 - V_1 + V_5$
0	0	1	0	1	13	$\Phi_{14} - \Phi_{13}$	$V_{10} - V_2 + V_5$
1	0	0	0	1	14	$\Phi_{15} - \Phi_{14}$	$V_{10} - V_1 + V_5$
0	0	0	0	1	15	$\pi - 2\Phi_{15}$	$V_{10} + V_5$
1	0	0	0	1	14	$\Phi_{15} - \Phi_{14}$	$V_{10} - V_1 + V_5$
0	0	1	0	1	13	$\Phi_{14} - \Phi_{13}$	$V_{10} - V_2 + V_5$
1	0	1	0	1	12	$\Phi_{13} - \Phi_{12}$	$V_{10} - V_2 - V_1 + V_5$
0	1	0	0	1	11	$\Phi_{12} - \Phi_{11}$	$V_{10} + V_1$
1	1	0	0	1	10	$\Phi_{11} - \Phi_{10}$	$V_{10}$
0	1	1	0	1	9	$\Phi_{10} - \Phi_9$	$V_{10} - V_2 + V_1$
1	1	1	0	1	8	$\Phi_9 - \Phi_8$	$V_{10} - V_2$
0	0	0	1	1	7	$\Phi_8 - \Phi_7$	$V_5 + V_2$
1	0	0	1	1	6	$\Phi_7 - \Phi_6$	$V_5 + V_2 - V_1$
0	0	1	1	1	5	$\Phi_6 - \Phi_5$	$V_5$
1	0	1	1	1	4	$\Phi_5 - \Phi_4$	$V_5 - V_1$
0	1	0	1	1	3	$\Phi_4 - \Phi_3$	$V_1 + V_2$
1	1	0	1	1	2	$\Phi_3 - \Phi_2$	$V_2$
0	1	1	1	1	1	$\Phi_2 - \Phi_1$	$V_1$
1	1	1	1	1	0	$\Phi_1$	No DC source

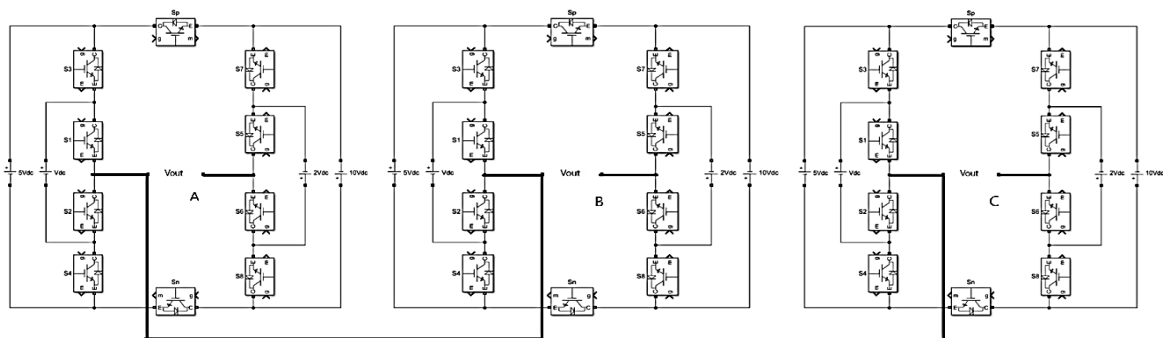


Figure 16. A proposed three-phase star-connected 31-level inverter



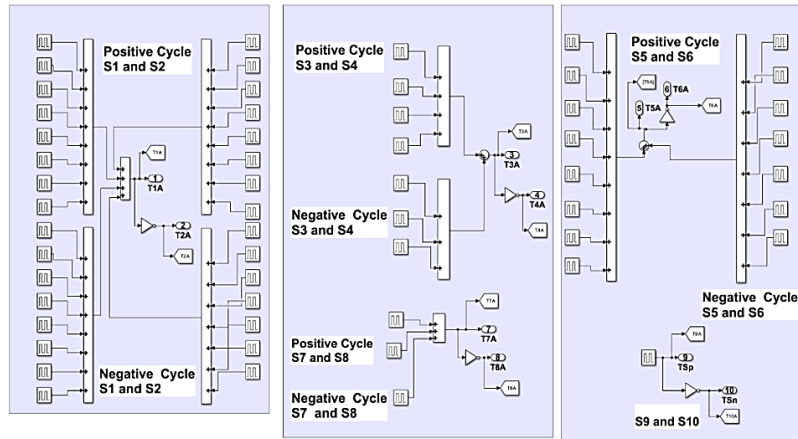


Figure 17. Simulink model for the switch gating of the positive half cycle of phase A

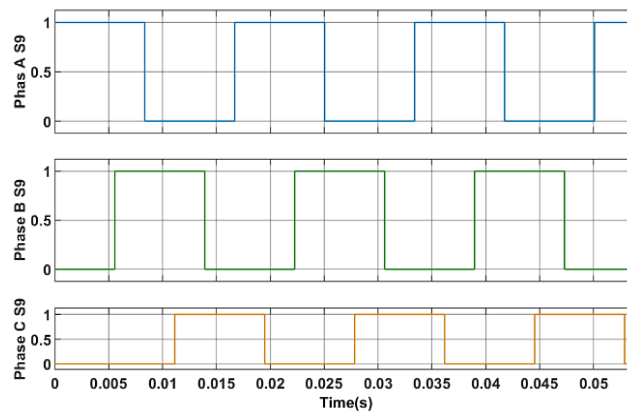


Figure 18. 120° phase difference between the three phases

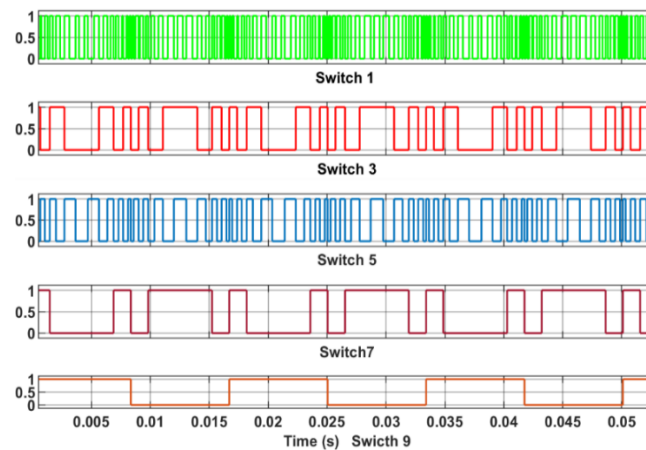


Figure 19. Gating angles for the ten power switches for phase A

Figure 20 shows the three-phase voltages, where the phase difference between the three phases is noted. Figure 21 shows the line voltages. The phase difference between the phase voltage and the line voltage is 30° as shown in Figure 22. The current drawn from the inductive load ( $R = 5 \Omega$ ,  $L = 20 \text{ mH}$ ) is illustrated in Figure 23. The current waveform is very close to a typical sine wave.

The phase difference between the phase voltage and current caused by the inductive load is shown in Figure 24. The neutral current  $I_n$  is very close to zero, which supports the precise operation of the three-phase inverter, as shown in Figure 25. To measure the THD of voltages and currents, the FFT tool in

MATLAB is used. Figure 26 shows  $\text{THD}\% = 4.71\%$  for the phase voltage,  $\text{THD}\% = 4.05\%$  for the line voltages are shown in Figure 27, and  $\text{THD}\% = 0.72\%$  for the load current is illustrated in Figure 28. These  $\text{THD}\%$  are considered low values according to the standard IEEE 519-2014 permissible values for the voltage range  $V \leq 1.0 \text{ kV}$  is  $12\%$ . The more angles, the lower values of  $\text{THD}\%$  are obtained.

To test the three-phase inverter for dynamic loads, the inverter is loaded with a three-phase asynchronous motor with  $460 \text{ V}$ ,  $60 \text{ Hz}$  specifications. To adjust the values of the DC sources to suit the characteristics of the motor the following DC sources are employed  $V_{\text{DC}} = 24 \text{ V}$ ,  $2 V_{\text{DC}} = 48 \text{ V}$ ,  $5 V_{\text{DC}} = 120 \text{ V}$ , and  $10 V_{\text{DC}} = 240 \text{ V}$ . Figures 29 and 30 show the phase and line voltages suitable to drive three-phase dynamic loads.

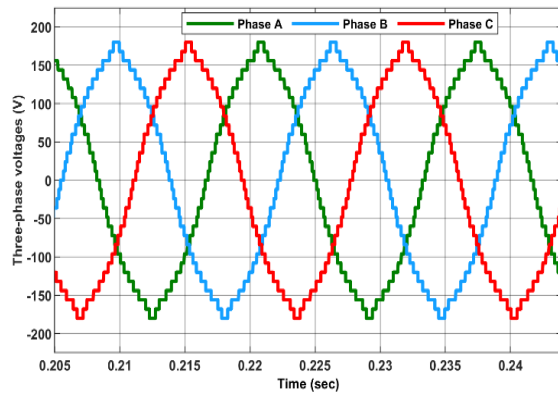


Figure 20. Three-phase voltages of the 31-level MLI

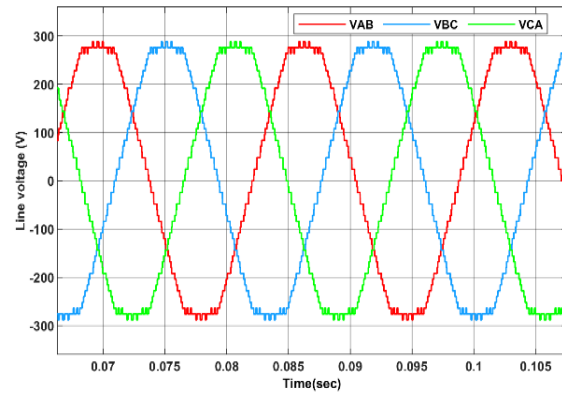


Figure 21. Three-line voltages of the 31-level MLI

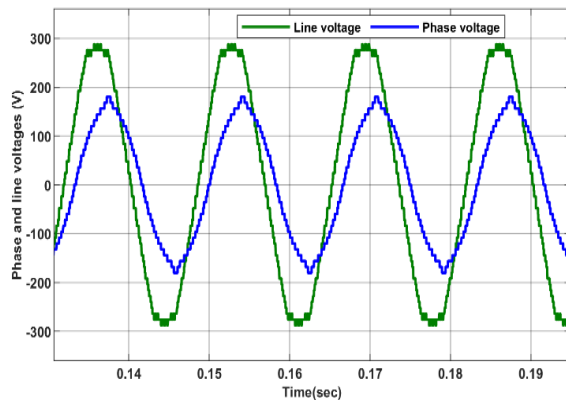


Figure 22. Phase difference phase-line voltages (RL load)

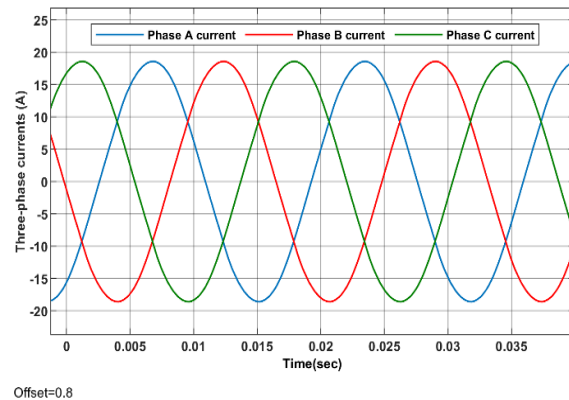


Figure 23. Three-phase current (RL load)

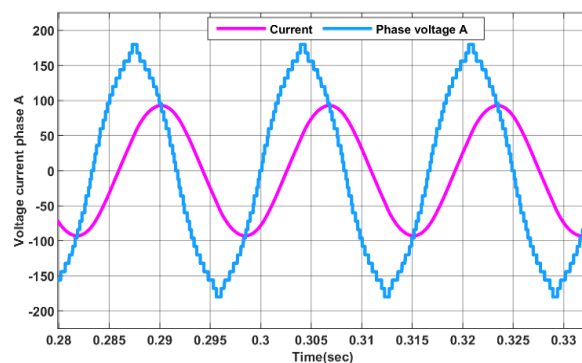


Figure 24. Phase difference voltage and current (RL load)

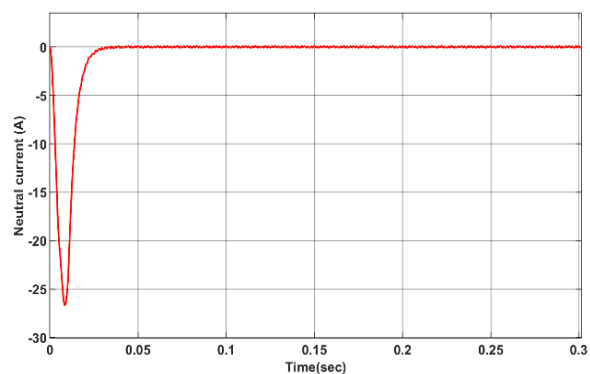


Figure 25. Neutral current  $I_n$



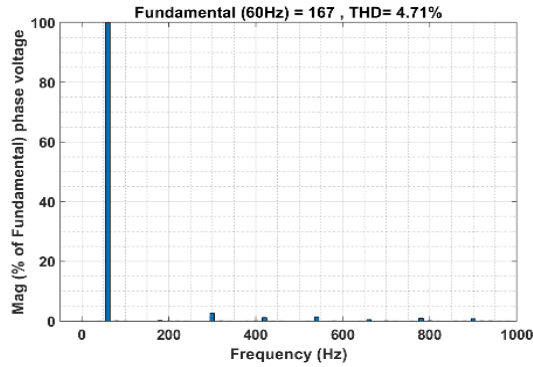


Figure 26. Phase voltage THD%

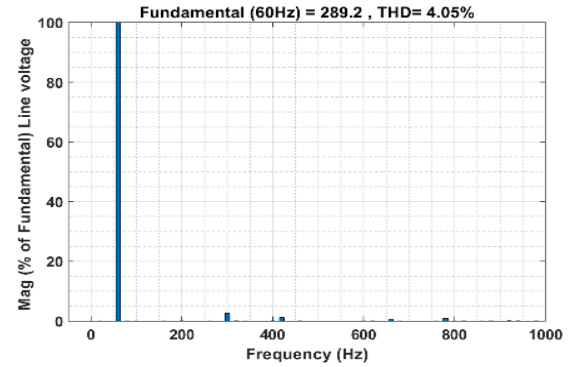


Figure 27. Line voltage THD%

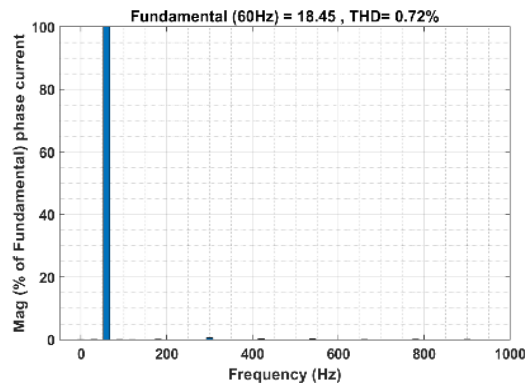


Figure 28. Load current THD% (RL)

The motor speed, electromagnetic torque, and current drawn are monitored. The motor is subjected to an external torque at 0.5 (sec) and 0.8 (sec) of the running time in succession. A fast response of all the mentioned variables is observed. Finally, the motor lost its speed at the timing 1 (sec) after being subjected to an external torque higher than the motor operating specification, as illustrated in Figure 31.

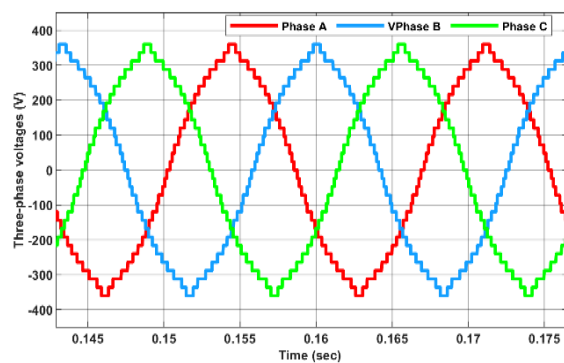


Figure 29. Three-phase 31-level MLI phase voltages

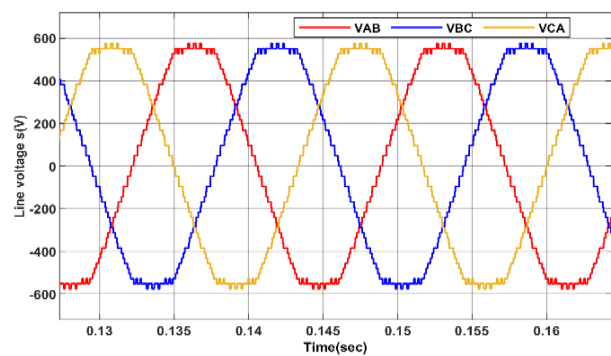


Figure 30. Three-line voltages 31-level MLI

### 3.2. FPGA simulation based on Spartan 3E controller

FPGA simulation utilizes the Spartan 3E controller as its processing unit. The controller operates at clock frequency of 50 MHz, resulting in a clock period of 20 ns. To complete a 60 Hz AC cycle supply at the inverter output, 833,333 clock cycles from the Spartan 3E controller are required. The gating of the thirty power transistor switches is programmed by employing the Spartan 3E by writing a program in VHDL for the 31-level three-phase MLI. Figure 32 shows the Spartan 3E simulation results, including clock frequency and the simulated pulses for the thirty power transistors. The VHDL file (UCF) indexing the addresses of the thirty output pins used

to gate the thirty power transistors is shown in Figure 33. These Spartan 3E thirty-pin can be connected to the gates of the power transistors via optocouplers to build an experimental setup for a 31-multi-level inverter.

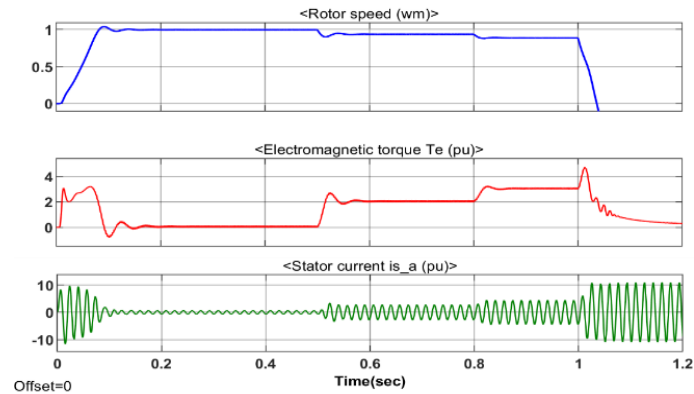


Figure 31. Rotor speed, electromagnetic torque, and stator current of the squirrel cage motor

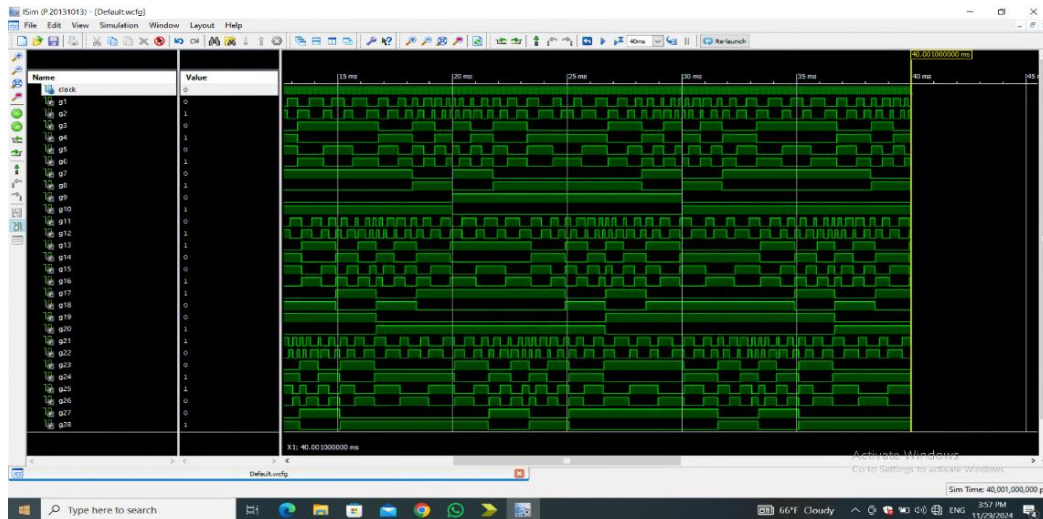


Figure 32. Clock frequency and the simulated pulses for the thirty power transistors

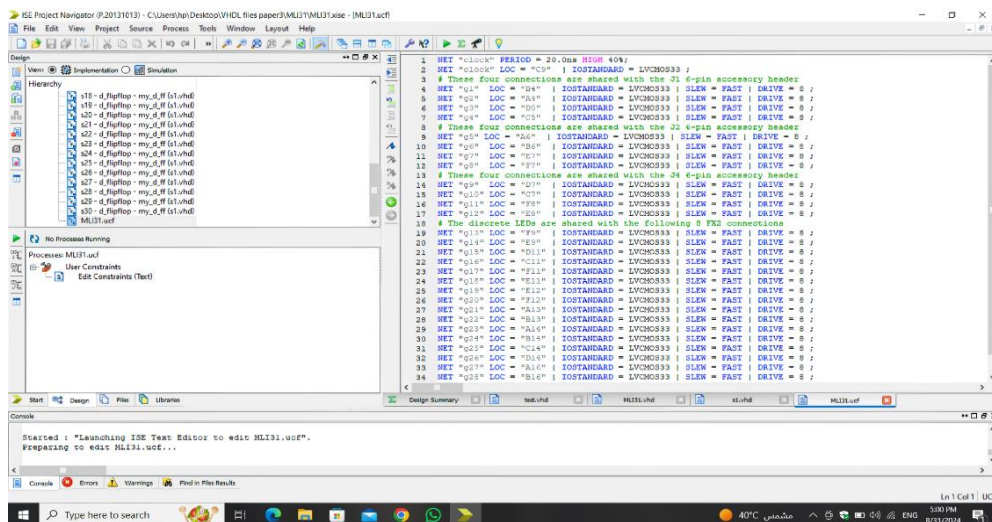


Figure 33. Addresses of the Spartan 3E output pins to gate thirty power transistors

#### 4. CONCLUSION

The first objective is the topological design of a 3-phase inverter to be used as a 3-phase voltage source. The goal is that the design comprises the minimal number of DC sources and the least number of power switches. After an extensive review of the research related to this objective, a 31-level three-phase is designed, where the required number of DC sources for each phase are 4 sources and the required number of power switches is 10, and in total 12 DC sources and 30 power switches for the three phases. The second objective is to lower the THD value of the MLI output waveform based on intelligent optimal algorithms. The characteristics of several intelligent optimal algorithms are reviewed in section 2 and five algorithms known for their extensive use by researchers are evaluated, namely, GA, GWO, PSO, SMA, and WOA, to achieve low THD% values over a wide modulation index range ( $0 < m < 1$ ). An optimal hybrid curve (OHC) is derived from the five algorithms i.e. GA, GWO, PSO, SMA, and WOA by taking the lowest value of the THD for a specific value of modulation index and is employed for the performance of 31-level MLI with static and dynamic loads. The third objective is to evaluate the efficiency of the proposed 31-level MLI integrated with intelligent optimization algorithms through model simulation. The thirty-one-level multi-level inverter is simulated by building a Simulink model in MATLAB using the optimal angles extracted from the hybrid optimization curve. A set of variables such as currents and voltages at the inverter output are monitored, and the frequency pattern analysis for different cases. A distinctive response to changes due to changing loads is observed, which makes the inverter reliable as a three-phase voltage source. The THD% are within 4% for inverter voltages and within 1% for the currents which is considered low values according to the IEEE standards.

#### ACKNOWLEDGEMENTS

The authors would like to acknowledge the support received from NTU/Iraq, USM/Malaysia, USM offshore/Türkiye, and PMU/Saudi Arabia.

#### FUNDING INFORMATION

Authors state no funding involved.

#### AUTHOR CONTRIBUTIONS STATEMENT

This journal uses the Contributor Roles Taxonomy (CRediT) to recognize individual author contributions, reduce authorship disputes, and facilitate collaboration.

Name of Author	C	M	So	Va	Fo	I	R	D	O	E	Vi	Su	P	Fu
Taha Ahmad Hussein	✓	✓		✓	✓	✓		✓	✓	✓				
Dahaman Ishak		✓		✓		✓		✓		✓	✓	✓		
Jawad Hasan	✓			✓		✓			✓		✓		✓	
Alkhateeb														
Mohamed Salem		✓	✓		✓			✓	✓				✓	

C : **C**onceptualization

M : **M**ethodology

So : **S**oftware

Va : **V**alidation

Fo : **F**ormal analysis

I : **I**nterpretation

R : **R**esources

D : **D**ata Curation

O : **O**rganizing - **O**rganizing

E : **E**ditorial - **E**ditorial

Vi : **V**isualization

Su : **S**upervision

P : **P**roject administration

Fu : **F**unding acquisition

#### CONFLICT OF INTEREST STATEMENT

Authors state no conflict of interest.

#### DATA AVAILABILITY

The authors confirm that the data supporting the findings of this study are available within the article.




#### REFERENCES

- [1] M. D. Siddique, S. Mekhilef, N. M. Shah, and M. A. Memon, "Optimal design of a new cascaded multilevel inverter topology with reduced switch count," *IEEE Access*, vol. 7, pp. 24498–24510, 2019, doi: 10.1109/ACCESS.2019.2890872.




- [2] P. Rajavel and V. J. Asokan, "A new configuration of multilevel inverter to generate higher voltage level with lower components," *International Journal of Power Electronics and Drive Systems*, vol. 15, no. 3, pp. 1508–1516, 2024, doi: 10.11591/ijpeds.v15.i3.pp1508-1516.
- [3] R. R. Kumar and J. Choudhary, "A novel multilevel inverter with reduced components and minimized voltage unbalance," *International Journal of Power Electronics and Drive Systems*, vol. 13, no. 4, pp. 2365–2377, 2022, doi: 10.11591/ijpeds.v13.i4.pp2365-2377.
- [4] P. V. V. R. Sharma and N. K., "A review on intelligence algorithms for THD minimization in multi-level inverter," *International Journal of Modelling and Simulation*, pp. 1–7, 2024, doi: 10.1080/02286203.2023.2298174.
- [5] A. Balal, S. Dinkhah, F. Shahabi, M. Herrera, and Y. L. Chuang, "A Review on multilevel inverter topologies," *Emerging Science Journal*, vol. 6, no. 1, pp. 185–200, Feb. 2022, doi: 10.28991/ESJ-2022-06-01-014.
- [6] M. H. Rashid, "Power electronics: devices, circuits, and applications," p. 998, 2014.
- [7] Z. Kang, Y. Guan, J. Wang, and P. Chen, "Research on genetic algorithm optimization with fusion Tabu search strategy and its application in solving three-dimensional packing problems," *Symmetry*, vol. 16, no. 4, p. 449, Apr. 2024, doi: 10.3390/sym16040449.
- [8] M. Neghină, A. I. Dicoiu, R. Chiş, and A. Florea, "A competitive new multi-objective optimization genetic algorithm based on apparent front ranking," *Engineering Applications of Artificial Intelligence*, vol. 132, 2024, doi: 10.1016/j.engappai.2024.107870.
- [9] G. G. Jin and A. K. Jarso, "An improved hybrid genetic algorithm using the affine combination-based reproduction," *Communications in Statistics - Simulation and Computation*, pp. 1–27, Jun. 2024, doi: 10.1080/03610918.2024.2363958.
- [10] E. Alkafaween, A. Hassanat, E. Essa, and S. Elmougy, "An efficiency boost for genetic algorithms: Initializing the GA with the iterative approximate method for optimizing the traveling salesman problem—experimental insights," *Applied Sciences*, vol. 14, no. 8, p. 3151, Apr. 2024, doi: 10.3390/app14083151.
- [11] M. Rostami, K. Berahmand, and S. Forouzanmehr, "A novel community detection based genetic algorithm for feature selection," *Journal of Big Data*, vol. 8, no. 1, 2021, doi: 10.1186/s40537-020-00398-3.
- [12] Y. Qiu, X. Yang, and S. Chen, "An improved gray wolf optimization algorithm solving to functional optimization and engineering design problems," *Scientific Reports*, vol. 14, no. 1, p. 14190, Jun. 2024, doi: 10.1038/s41598-024-64526-2.
- [13] G. Shial, S. Sahoo, and S. Panigrahi, "An enhanced GWO algorithm with improved explorative search capability for global optimization and data clustering," *Applied Artificial Intelligence*, vol. 37, no. 1, Dec. 2023, doi: 10.1080/08839514.2023.2166232.
- [14] J.-S. Wang and S.-X. Li, "An improved grey wolf optimizer based on differential evolution and elimination mechanism," *Scientific Reports*, vol. 9, no. 1, p. 7181, May 2019, doi: 10.1038/s41598-019-43546-3.
- [15] N. Singh and S. B. Singh, "A novel hybrid GWO-SCA approach for optimization problems," *Engineering Science and Technology, an International Journal*, vol. 20, no. 6, pp. 1586–1601, Dec. 2017, doi: 10.1016/j.jestech.2017.11.001.
- [16] R. A. Khanum, M. A. Jan, A. Aldegheshem, A. Mehmood, N. Alrajeh, and A. Khanan, "Two new improved variants of grey wolf optimizer for unconstrained optimization," *IEEE Access*, vol. 8, pp. 30805–30825, 2020, doi: 10.1109/ACCESS.2019.2958288.
- [17] J. Kennedy and R. Eberhart, "Particle swarm optimization," in *Proceedings of ICNN'95 - International Conference on Neural Networks*, IEEE, 1995, pp. 1942–1948. doi: 10.1109/ICNN.1995.488968.
- [18] X. Zhang and Y. Yang, "Optimization of PID controller parameters using a hybrid PSO algorithm," *International Journal of Dynamics and Control*, vol. 12, no. 10, pp. 3617–3627, Oct. 2024, doi: 10.1007/s40435-024-01455-y.
- [19] B. Cortés-Cáicedo, J. Ocampo-Toro, R. I. Bolaños, O. D. Montoya, and L. F. Grisales-Noreña, "A multi-objective PSO for DC microgrids: Efficient battery management to minimize energy losses and operating costs," *Journal of Energy Storage*, vol. 96, p. 112550, Aug. 2024, doi: 10.1016/j.est.2024.112550.
- [20] S. Sudha Letha, T. Thakur, and J. Kumar, "Harmonic elimination of a photo-voltaic based cascaded H-bridge multilevel inverter using PSO (particle swarm optimization) for induction motor drive," *Energy*, vol. 107, pp. 335–346, 2016, doi: 10.1016/j.energy.2016.04.033.
- [21] R. Djidimbelé, B.-P. Ngoussandou, D. K. Kidmo, Kitmo, M. Bajaj, and D. Raidandi, "Optimal sizing of hybrid systems for power loss reduction and voltage improvement using PSO algorithm: Case study of Guissia rural grid," *Energy Reports*, vol. 8, pp. 86–95, Nov. 2022, doi: 10.1016/j.egy.2022.06.093.
- [22] S. Murugesan and M. V. Suganyadevi, "Performance analysis of simplified seven-level inverter using hybrid HHO-PSO algorithm for renewable energy applications," *Iranian Journal of Science and Technology - Transactions of Electrical Engineering*, vol. 48, no. 2, pp. 781–801, 2024, doi: 10.1007/s40998-023-00676-9.
- [23] S. Khunkitti, A. Siritatiwat, and S. Premrudeepreechacharn, "Multi-objective optimal power flow problems based on slime mould algorithm," *Sustainability (Switzerland)*, vol. 13, no. 13, 2021, doi: 10.3390/su13137448.
- [24] V. Malathy, J. N. Kalshetty, K. Al-Attabi, M. R. Kamesh, and A. Das, "Optimization of renewable energy using hybrid multi-objective slime mould and simulated annealing algorithms," in *2024 Second International Conference on Data Science and Information System (ICDSIS)*, IEEE, May 2024, pp. 1–5, doi: 10.1109/ICDSIS61070.2024.10594670.
- [25] M. Mostafa, H. Rezk, M. Aly, and E. M. Ahmed, "A new strategy based on slime mould algorithm to extract the optimal model parameters of solar PV panel," *Sustainable Energy Technologies and Assessments*, vol. 42, 2020, doi: 10.1016/j.seta.2020.100849.
- [26] K. Yu, L. Liu, and Z. Chen, "An improved slime mould algorithm for demand estimation of urban water resources," *Mathematics*, vol. 9, no. 12, p. 1316, Jun. 2021, doi: 10.3390/math9121316.
- [27] F. S. Gharehchopogh, A. Ucan, T. Ibrici, B. Arasteh, and G. Isik, "Slime mould algorithm: A comprehensive survey of its variants and applications," *Archives of Computational Methods in Engineering*, vol. 30, no. 4, pp. 2683–2723, 2023, doi: 10.1007/s11831-023-09883-3.
- [28] Y. Wei, Z. Othman, K. M. Daud, Q. Luo, and Y. Zhou, "Advances in slime mould algorithm: a comprehensive survey," *Biomimetics*, vol. 9, no. 1, p. 31, Jan. 2024, doi: 10.3390/biomimetics9010031.
- [29] S. Mirjalili and A. Lewis, "The whale optimization algorithm," *Advances in Engineering Software*, vol. 95, pp. 51–67, May 2016, doi: 10.1016/j.advengsoft.2016.01.008.
- [30] F. S. Gharehchopogh and H. Gholizadeh, "A comprehensive survey: Whale optimization algorithm and its applications," *Swarm and Evolutionary Computation*, vol. 48, pp. 1–24, Aug. 2019, doi: 10.1016/j.swevo.2019.03.004.
- [31] S. Chakraborty, A. Kumar Saha, S. Sharma, S. Mirjalili, and R. Chakraborty, "A novel enhanced whale optimization algorithm for global optimization," *Computers & Industrial Engineering*, vol. 153, p. 107086, Mar. 2021, doi: 10.1016/j.cie.2020.107086.
- [32] Q.-V. Pham, S. Mirjalili, N. Kumar, M. Alazab, and W.-J. Hwang, "Whale optimization algorithm with applications to resource allocation in wireless networks," *IEEE Transactions on Vehicular Technology*, vol. 69, no. 4, pp. 4285–4297, 2020, doi: 10.1109/TVT.2020.2973294.
- [33] M. H. Nadimi-Shahraki, H. Zamani, Z. Asghari Varzaneh, and S. Mirjalili, "A systematic review of the whale optimization algorithm: Theoretical foundation, improvements, and hybridizations," *Archives of Computational Methods in Engineering*, vol. 30, no. 7, pp. 4113–4159, Sep. 2023, doi: 10.1007/s11831-023-09928-7.

## BIOGRAPHIES OF AUTHORS






**Taha Ahmed Hussein**    is currently an assistant professor at the Department of Electrical Engineering, Northern Technical University Mosul, Iraq. He received his B.Sc. and M.Sc. degrees in Electrical Engineering from University of Technology and Mosul University Iraq, in 1982 and 1993, respectively. His research interests include the fields of power electronics, control engineering, and optimization algorithms. He can be contacted at email: taha.hussien@ntu.edu.iq.






**Dahaman Ishak**    is currently an associate professor at the Department of Electrical Engineering, Prince Mohammad Bin Fahd University, Al Khobar 31952, Saudi Arabia. He received a B.Sc. degree in Electrical Engineering from Syracuse University, Syracuse, NY, USA, an M.Sc. degree in Electrical Power from the University of Newcastle Upon Tyne, U.K., and a Ph.D. degree from the University of Sheffield, U.K., in 1990, 2001, and 2005, respectively. His current research interests include electrical machines and drives, power electronic converters, and renewable energy. He can be contacted at email: dbinishak@pmu.edu.sa.



**Jawad Hasan Alkhateeb**    is currently an assistant professor in the Computer Engineering Department at Prince Mohammad Bin Fahd University, Saudi Arabia. He received B.Sc. degree in Electrical Engineering from Louisiana Tech University, USA in 1991, M.Sc. degree in Electrical and Computer Engineering from Louisiana Tech University, USA in 1992, and Ph.D. in Computer Science and Engineering from the University of Bradford, UK, in 2010. He has 28 years of academic experience in teaching computer science and engineering topics. His research interests include AI, machine learning, deep learning, pattern recognition, digital image processing, and IoT. He can be contacted at email: jalkhateeb@pmu.edu.sa.



**Mohamed Salem**    received a B.Eng. in Electrical and Power Engineering from Elmergib University, Al Khums, Libya, in 2008. The M.Sc. in Electrical Engineering from Tun Hussein Onn University of Malaysia (UTHM), Malaysia, in 2011. In August 2017, He was awarded his Ph.D. degree from Universiti Teknologi Malaysia (UTM), Malaysia. He is a member and a registered graduate engineer in Malaysia (BEM) in the electrical track. Currently, he is a senior lecturer at the School of Electrical and Electronic Engineering, Universiti Sains Malaysia, Malaysia since July 2018. Since 2022, he has been appointed as a researcher at Libyan Authority for Scientific Research. His research interests are in DC-DC converters, renewable energy applications, energy conversion, and control of power electronics systems. He can be contacted at email: salemm@usm.my.